



Linear Algebra Review Part II

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Inverse

- The inverse of matrix A is denoted as A^{-1} .
- If matrix A has an inverse, then it should be full rank or in other words, its determinant is non-zero.
- $(A^{-1})^{-1} = A$
- $(AB)^{-1} = B^{-1}A^{-1}$
- $(A^{-1})^T = (A^T)^{-1}$

Trace

- The trace of a square matrix A is sum of the diagonal elements.
- $\text{tr}A = \text{tr}A^T$
- $\text{tr}(A + B) = \text{tr}A + \text{tr}B$
- $\text{tr}(tA) = t \text{tr}A$
- For matrices A, B such that AB is a square matrix, then
$$\text{tr}(AB) = \text{tr}(BA)$$
- For matrices A, B, C such that ABC is a square matrix, then
$$\text{tr}(ABC) = \text{tr}(BCA) = \text{tr}(CAB)$$

Linearly Independent and Dependent

- For a set of vectors $\{x_1, x_2, \dots, x_n\}$, they are said to be linearly independent if no vector can be represented as a linear combination of the remaining vectors.
- Conversely, if one vector belonging to the set can be represented as a linear combination of the remaining vectors, then the vectors are said to be linearly dependent. In this case,

$$x_n = a_1 x_1 + a_2 x_2 + \dots + a_{n-1} x_{n-1}$$

Rank

- The column rank of matrix A is the size of the largest subset of columns of A that constitute a linearly independent set. Likewise, the row rank of matrix A is the size of the largest subset of columns of A that constitute a linearly independent set.
- The column rank of a matrix is equal to the row rank of a matrix, so their value is called the rank of a matrix.

Rank

- For an $n \times m$ matrix \mathbf{A} , $\text{rank}(\mathbf{A}) \leq \min(n, m)$ and if $\text{rank}(\mathbf{A}) = \min(n, m)$, then matrix \mathbf{A} is said to be full rank.
- $\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{A}^T)$
- $\text{rank}(\mathbf{A} + \mathbf{B}) \leq \text{rank}(\mathbf{A}) + \text{rank}(\mathbf{B})$
- For an $n \times m$ matrix \mathbf{A} and a $m \times p$ matrix \mathbf{B} ,
$$\text{rank}(\mathbf{AB}) \leq \min(\text{rank}(\mathbf{A}), \text{rank}(\mathbf{B}))$$

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Homogeneous Linear Equation

- $A X = 0$ is called homogeneous linear equation.
- Steps of determinate the structure of the solution:
 1. Calculate the determinant of A
 2. If the result is non-zero, then there is only zero solution;
otherwise, there is an infinite number of non-zero solutions.

Non-homogeneous Linear Equation

- $A X = b$ (where b is not a zero-vector) is called non-homogeneous linear equation.
- Steps of determinate the structure of the solution:
 1. Calculate the rank of A
 2. Calculate the rank of A 's augmented matrix (putting vector b to the right of A to form this matrix), named B

Non-homogeneous Linear Equation

- There are three cases:
- If $\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{B}) = n$, then there is only one solution.
- If $\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{B}) < n$, then there is an infinite number of solutions.
- If $\text{rank}(\mathbf{A}) < \text{rank}(\mathbf{B})$, then there is no solution.



Thanks for
listening!